

Aim:- to study different activation functions used in deep learning implement them and analyze their role in introducing non-linearity improving learning capability

Algorithm:-

- Start
- Initialize the input values
- Define commonly used activation functions:
 - Sigmoid : $f(x) = \frac{1}{1+e^{-x}}$
 - Tanh : $f(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$
 - ReLU (rectified linear unit) :
 $f(x) = \max(0, x)$
 - Soft max : $f(x_i) = \frac{e^{x_i}}{\sum_j e^{x_j}}$

4, Apply each activation function on input data
 5, Plot/observe the output behaviour for different range of input

6, Train a small neural network with each activation function and compare performance

7, Stop

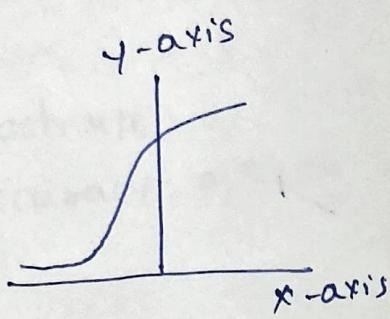
Pseudo code:

- start
- Load MNIST dataset
- Normalize image and flatten to vectors
- For each activation in [Sigmoid, tanh, relu]:
 - Build model with 2 hidden layers using that activation.
 - Train model for 5 epochs
 - Evaluate on test data.

~~TOP~~

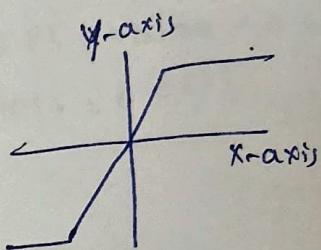
Sigmoid:

$$g(x) = \frac{1}{1+e^{-x}}$$

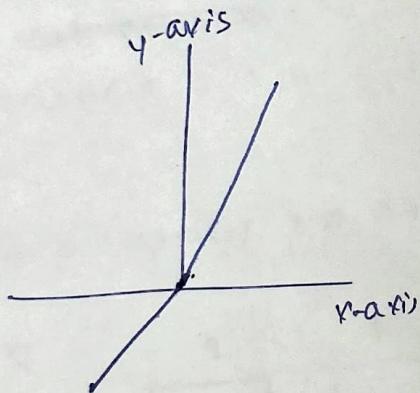


Tanh:

$$g(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

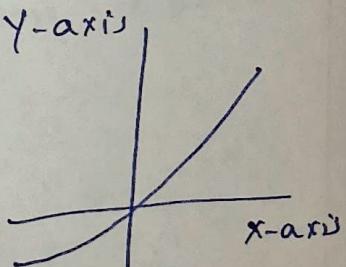


leaky relu:

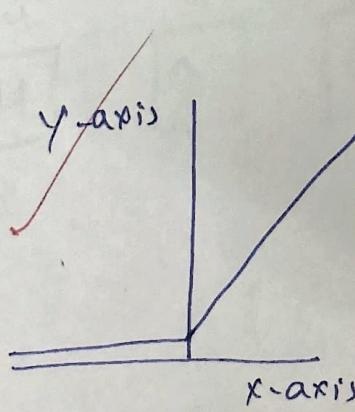


$$g(x) = \max(0, \alpha x, x)$$

Exponential linear units:



Relu:



$$g(x) = \max(0, x)$$

- Print accuracy and loss

END

observation:

- 1, the sigmoid activation function showed slower convergence and achieved lower accuracy compared to others
- 2, tanh performed better than sigmoid since it is zero centered, but still suffered from vanishing gradients
- 3, the choice of activation function had a significant effect on both training speed and final model accuracy

result: if successfully implemented the activation functions used in deep learning.

~~eff. 12319~~

```
import tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.datasets import mnist
from tensorflow.keras.utils import to_categorical

(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_train = x_train.reshape(-1, 784).astype('float32') / 255.0
x_test = x_test.reshape(-1, 784).astype('float32') / 255.0
y_train = to_categorical(y_train, 10)
y_test = to_categorical(y_test, 10)

for act in ['sigmoid', 'tanh', 'relu', 'elu']:
    model = models.Sequential([
        layers.Dense(128, activation=act, input_shape=(784,)),
        layers.Dense(64, activation=act),
        layers.Dense(10, activation='softmax')
    ])
    model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
    model.fit(x_train, y_train, epochs=5, batch_size=128, verbose=0)
    _, acc = model.evaluate(x_test, y_test, verbose=0)
    print(f"Activation: {act} | Test Accuracy: {acc:.4f}")
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>

```
11490434/11490434 0s 0us/step
/usr/local/lib/python3.12/dist-packages/keras/src/layers/core/dense.py:93: UserWarning: Do not pass an `input_shape`/`input_dim` argument to `Dense`. Instead, use `activity_regularizer=activity_regularizer` in the constructor.
  super().__init__(activity_regularizer=activity_regularizer, **kwargs)
Activation: sigmoid | Test Accuracy: 0.9609
Activation: tanh    | Test Accuracy: 0.9733
Activation: relu    | Test Accuracy: 0.9746
Activation: elu     | Test Accuracy: 0.9740
```